ECE 5256 Digital Image Processing S. Kozaitis

Chapter 6.10 Wavelet Transforms

Basis Functions of Other Transforms



FIGURE 6.3

Basis vectors
(for N = 16) of
some commonly
encountered
transforms:
(a) Fourier basis
(real and imaginary parts),
(b) discrete
Cosine basis,
(c) Walsh-Hadamard basis,
(d) Slant basis,
(e) Haar basis,
(f) Daubechies

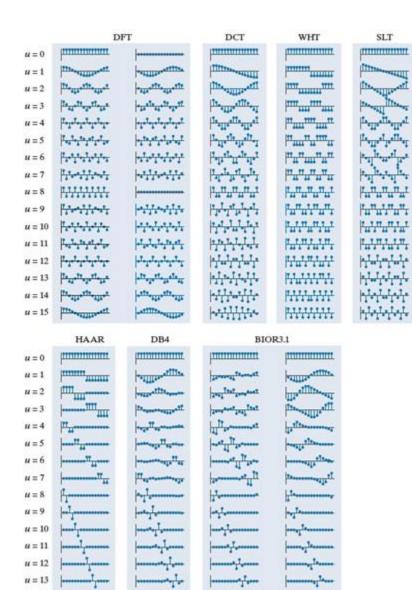
(g) Biorthogonal

its dual, and

B-spline basis and

u = 15

basis.

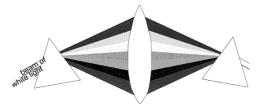


Bottom row are wavelet transforms

What is the main difference between the rows?

Wavelet Transforms are Similar to Fourier Transform But Often Better

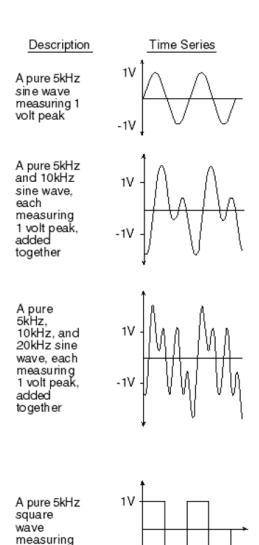
• A wavelet transform is a tool for analyzing data of different frequency ranges, allowing one to study each component separately.



 The basic idea of the wavelet transform is to represent any arbitrary function as a superposition of a set of such wavelets (basis functions). It is similar in concept to the Fourier Transform.

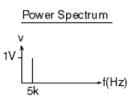
 The basis functions in the wavelet transform are very different from the Fourier transform. They are obtained from a single localized function which is scaled and shifted.

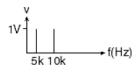
Fourier Transform Has Been Used for Frequency Analysis

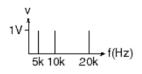


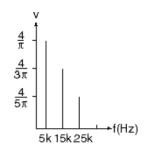
-1V

1 volt



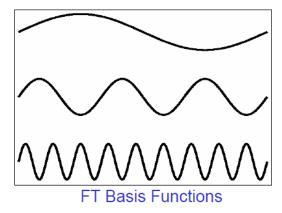






Fourier Transform Basis Functions are Sines and Cosines

- The Fourier transform (FT) represents a signal in terms of different frequencies
- Sines and cosines are the basis function of the Fourier transform.



 The basis functions of the Fourier transform are not localized (they extend to infinity

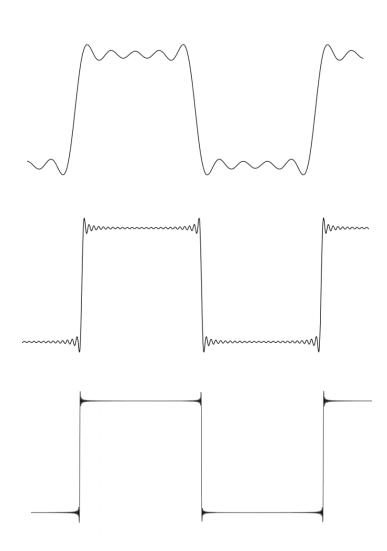
Many Basis Functions Needed to Reconstruct Edges (not good)

Square wave reconstructed with

5 harmonics

25 harmonics

125 harmonics

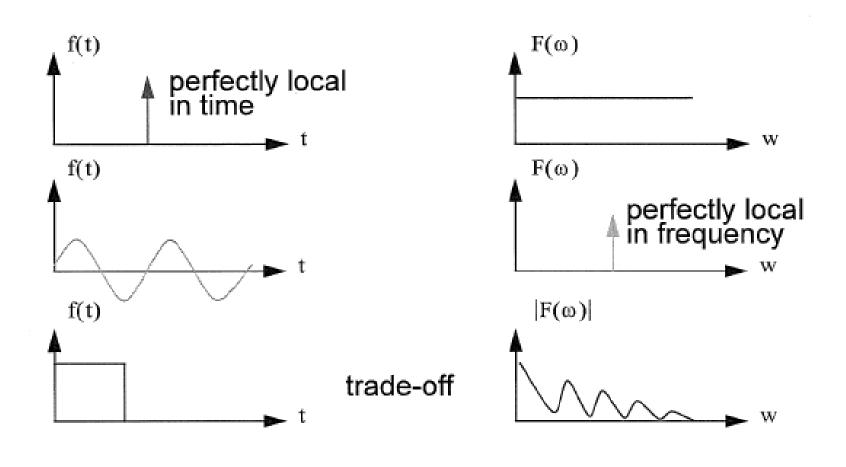


What is Wrong with the Fourier Transform?

- Short signals are described by many frequencies
- Edges, changes, beginning and ends of events are often difficult to isolate

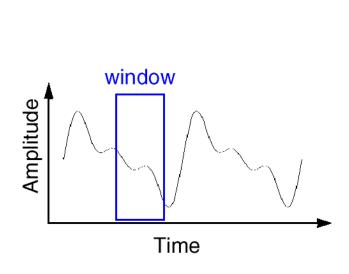
$$X(f) = \int_{-\infty}^{\infty} x(t)e^{-j2\pi jt}dt$$
basis functions

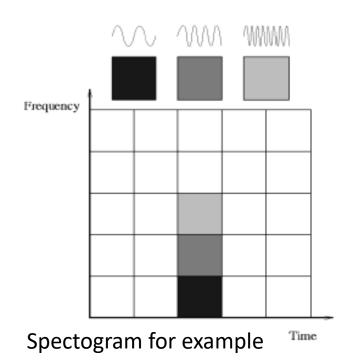
Trade-off between Time and Frequency Localization



Short Time Fourier Analysis Might Be Better

- Take the Fourier Transform of a "window" of a signal
- Both time and frequency are represented in limited precision





What is Wrong with the Short-Time Fourier Transform?

$$STFT_{x}(t,f) = \int [x(t')g(t'-t)]e^{-j2\pi ft'}dt'$$

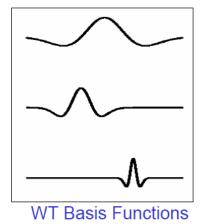
- Requires short time window (large frequency window) to extract features of a signal.
- Requires *short frequency* window (long time window) to extract features of a signal.

 Many signals require a more flexible approach – a variable window size to determine either time of frequency feature more accurately

What Are Wavelets?

Wavelets are localized waves that can be scaled and shifted

• Any signal can be decomposed in terms of wavelets.



Wavelets are the basis functions of a wavelet transforms

There are many wavelet transforms

Wavelet History

- 1805 Fourier analysis developed
- 1910 Alfred Haar discovers the Haar transform
- 1980's beginnings of wavelets in physics, vision, speech processing (ad hoc)
- … little theory … why/when do wavelets work?
- 1987 Mallat developed multiresolution theory, DWT, wavelet construction techniques (but still noncompact)
- 1988 Daubechies added theory: found compact, orthogonal wavelets with arbitrary number of vanishing moments!
- 1990's: wavelets took off, attracting both theoreticians and engineers

Comparison to Fourier Analysis

Fourier analysis

- Basis is global
- Sinusoids with frequencies in arithmetic progression

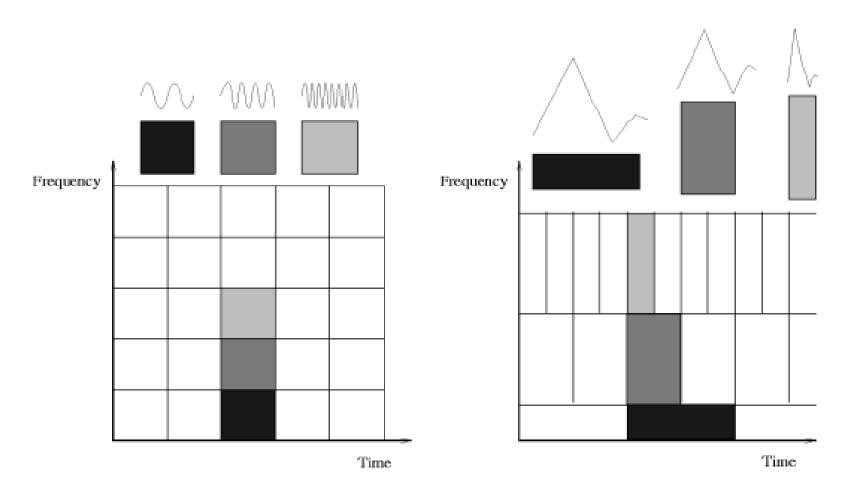
Short-time Fourier Transform

- Basis is local
- Fixed-width Gaussian "window"

Wavelet

- Basis is local
- Frequencies in geometric progression
- Basis has constant shape independent of scale

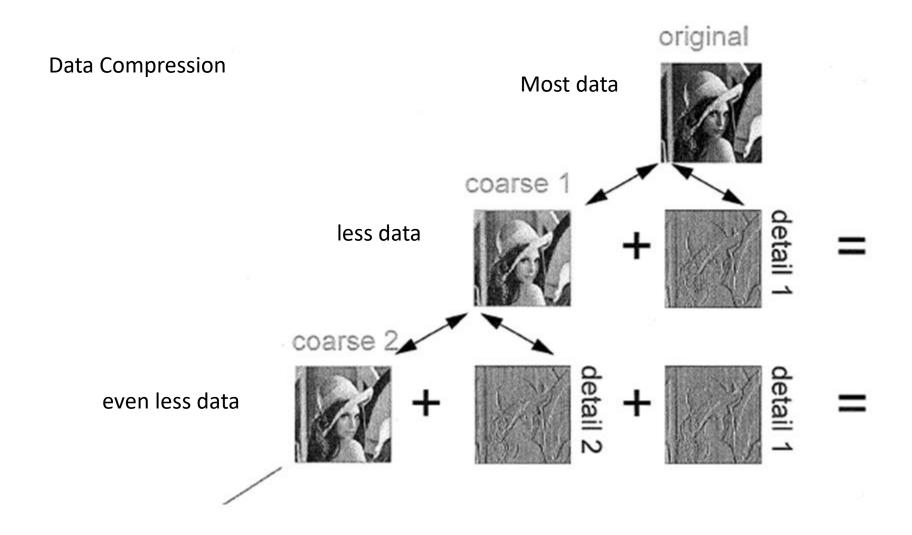
Wavelet Transform Automatically Produces a Multiresolution Analysis



Short-time Fourier Transform

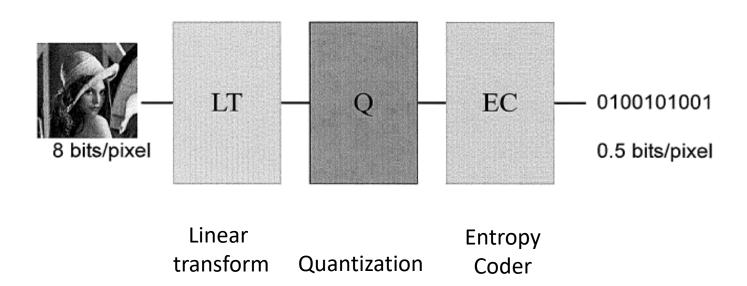
Wavelet Transform

Multiresolution Analysis is Useful for Refinement of a Signal



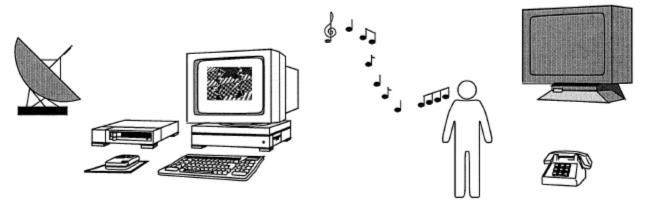
Wavelet Transform Compacts Energy Better Than the Fourier Transform

Data compression



Many Uses of Multiresolution

- digital audio and video coding
- · conversions between TV standards
- digital HDTV and audio broadcast
- remote image databases with searching
- storage media with random access
- MR coding for multicast over the Internet
- MR graphics



Compression: still a key technique in communications

Only One General Type of Useful Wavelet Transform

Continuous – useful for theoretical work
 Input continuous, wavelet continuous, shift and scales continuous

Other versions

- Input continuous, wavelet continuous, shift and scales discrete
- Input discrete, wavelet discrete, shift and scales discrete
- Discrete Dyadic (scales are powers of 2)
 - Input and wavelet are discrete and wavelets scale and shift by powers of two
 - N input samples and N output samples

Main Wavelet Transform Equation

$$b[j,k] = \sum_{n=0}^{n=N-1} f[n] 2^{j/2} \psi^*[2^j n - k]$$
 wavelet coefficients

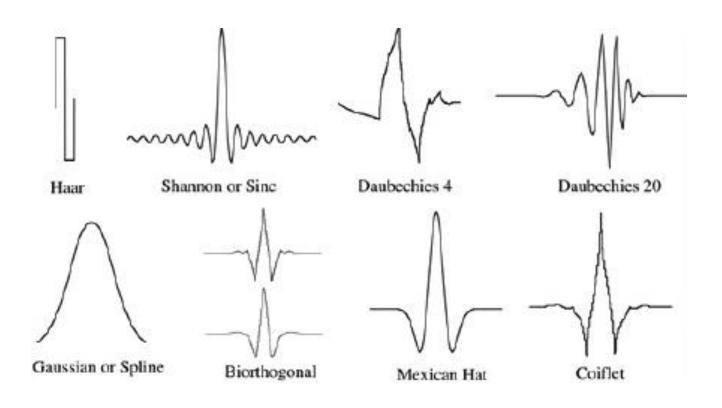
N = length of signal

 $j = 0, 1, 2 \dots$ (log₂N) -1 scales (0 scale is largest wavelet)

2^j shifts at each scale (Wavelet transform has j levels)

Looks complicated but easy implementation

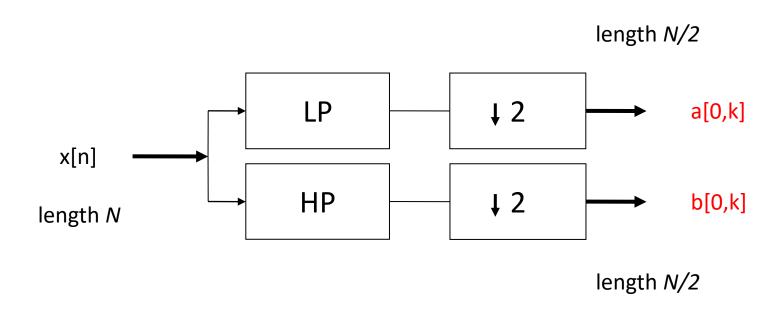
Infinite Number of Wavelets



Wavelets Implemented by Digital Filter Banks

(wavelet not obvious)

One level of the wavelet transform (level 0)



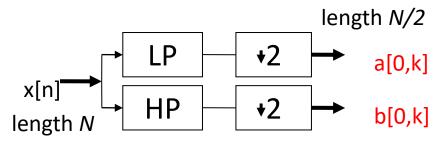
1-level of the WT is y[n] = [a[0,k] b[0,k]]

LP – low-pass filter

HP – high-pass filter

Wavelets Implemented by Digital Filter Banks

One level of the wavelet transform



length N/2

vectors

Filter coefficients = [f[0], f[1], ...]

Filter bank written in matrix form

$$[y] = \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

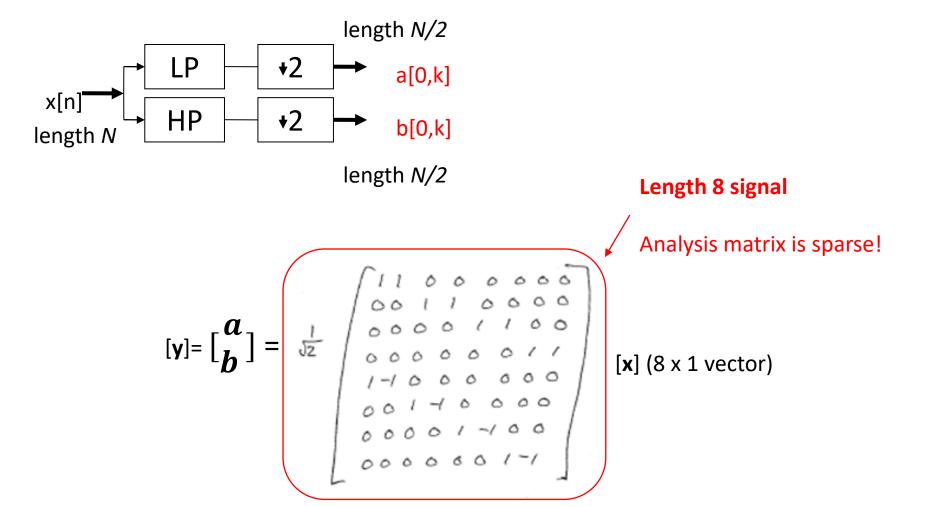
[x] (4 x 1 column vector)

The Haar wavelet is often used in examples because it is the shortest

- deconsampling removes every other

Wavelets Implemented by Digital Filter Banks

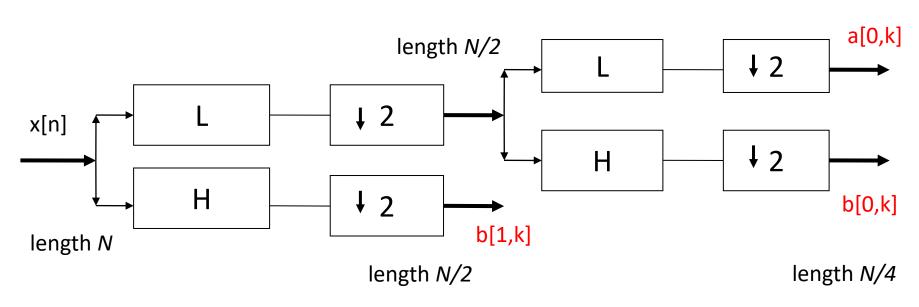
One level of the wavelet transform



Add a Filter Bank for Another Level

Add to the low-pass side for every level

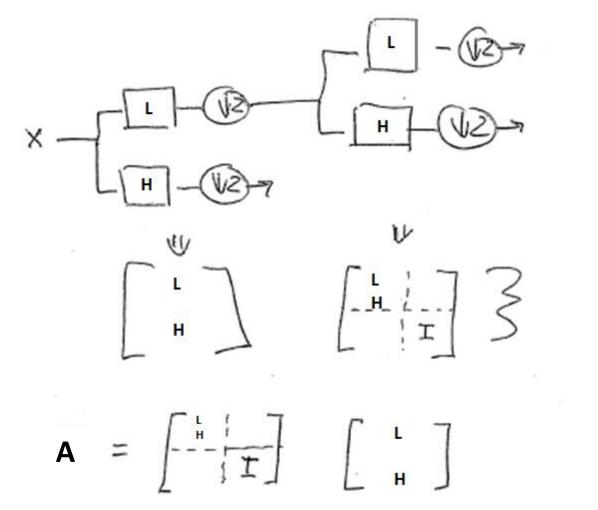




wavelet transform is always length N

2-levels of the WT is y[n] = [a[0,k] b[0,k] b[1,k]]

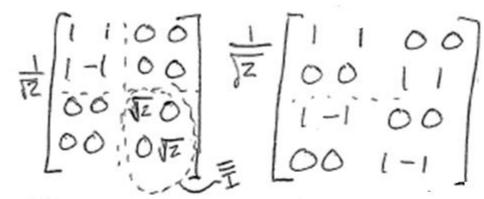
Generate the Analysis Matrix for Another Level

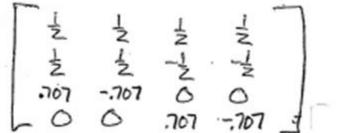


Multiply in reverse order of the block diagram

Add a Filter Bank for Another Level

$$\mathbf{A} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$



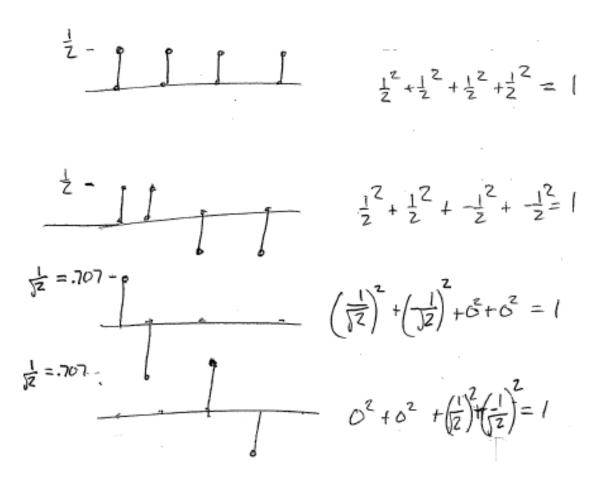


The rows of the **A** matrix are the basis functions of the transform - wavelets

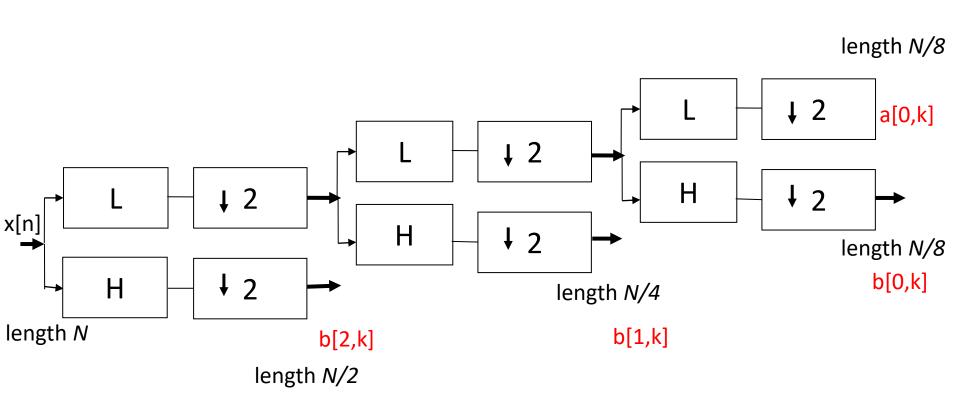
The first row is called a scaling function because it didn't use a high-pass filter, **a** coefficients are called scaling function coefficients

Multiplying a signal **x** by **A** is equivalent to correlating **x** with each basis function

- The four basis function (wavelets) are the rows of the A matrix
- They all have the same energy



N Banks for N Levels



Analysis Matrix of 3-level Haar Wavelet Transform for Length 8 signal

Easy to generate matrix from filter banks

Each matrix is 8 x 8

j is the level of the transform

The rows of the matrix are the wavelets!

Can Have Different Levels of a Wavelet Transform

These are all wavelet transforms of a length N signal

a[0,k] b[0,k]

Length [
$$N/2 + N/2$$
]

a[0,k] b[0,k] b[1,k]

Length [$N/4 + N/4 + N/2$]

a[0,k] b[0,k] b[1,k] b[2,k]

Length [$N/8 + N/8 + N/4 + N/2$]